AFRL-RH-WP-TR-2012-0089 Quantitative Structure-Activity Relationships for Organophosphate Enzyme Inhibition (Briefing Charts)

22 September 11

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14. ABSTRACT				
Organophosphates (OPs) are a group of	of pesticides that inhibit enzymes such as acetylcholines			
	to quickly obtain. To address this concern, quantitative			
	tylcholinesterase, butyrycholinesterase, trypsin and chy			
	DR) models. The acetylcholinesterase database consists			
	CODESSA descriptors (SemiChem, Inc.) were calculate the average nucleophilic reactive index for a carbon at			
acceptendiniesterase results show that	the average nucleophine reactive index for a carbon att	on contributed most significantly to uniding.		

A training R2 of 0.73±0.01 and an external test set Q2 of 0.62±0.06 was achieved. The QSAR models discussed in this seminar will complement OP BBDR modeling by filling critical data gaps for key parameter values, leading to better risk assessment and prioritization of animal and human toxicity studies especially for OPs lacking experimental data.

15. SUBJECT TERM	MS				
QSAR C	Organophosphates	Acetylcholinest	erase BBDR		
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Outline



1. Introduction

- a) What is QSAR?
- b) Organophosphate structure and mechanism of toxicity
- c) Linking QSAR and OP PBPK/PD

2. Methods

- a) Physiochemical Descriptors
- b) Regression Techniques

3. Results

- a) Bimolecular rate constants
 - i. Acetylcholinesterase & Butyrylcholinesterase
 - ii. Trypsin & Chymotrypsin

4. Discussion



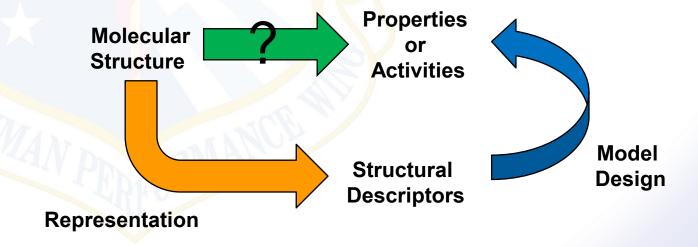


What is QSAR?



Quantitative Structure-Activity Relationship

A technique used to quantify differences between biological activity and that of a molecular structure



There are guidelines/rules to this approach

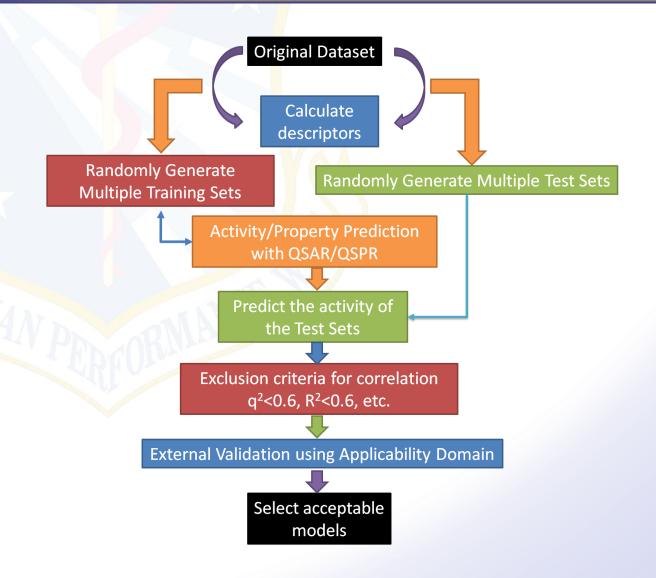
- 1. Choose well-defined Activity endpoints
- 2. Choose plausible molecular descriptors
- 3. Explore the data with statistics
- 4. Test hypotheses with new data (ie. iterate)





QSAR Overview





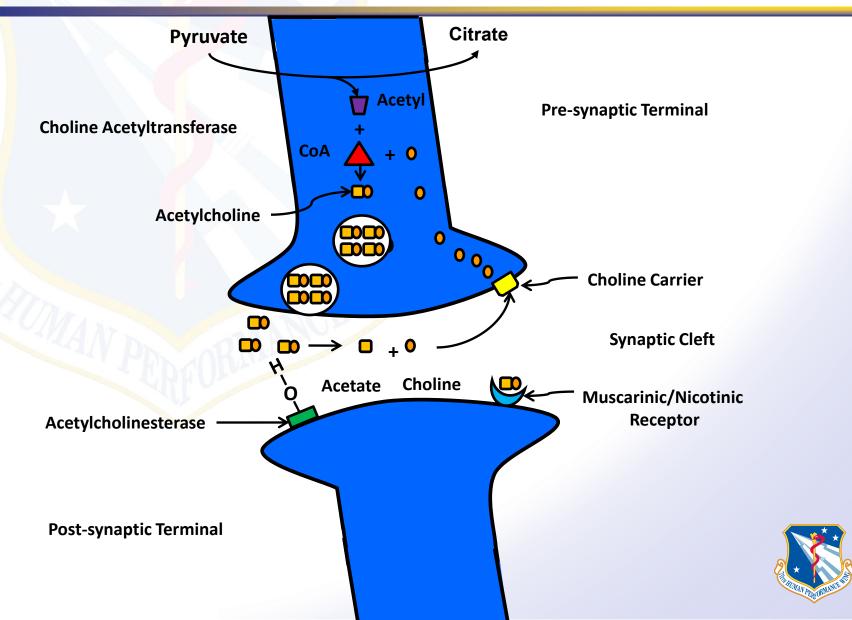




Cholinergic Nervous System



"Normal Mechanism of Action"

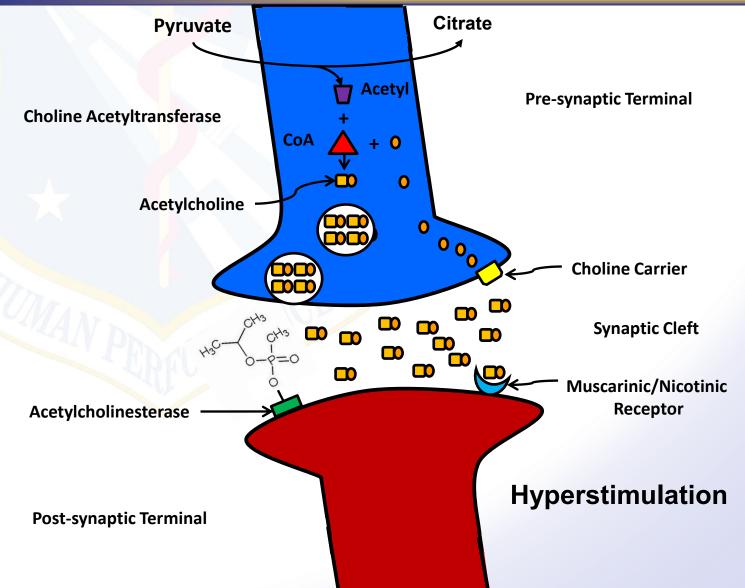




Cholinergic Nervous System



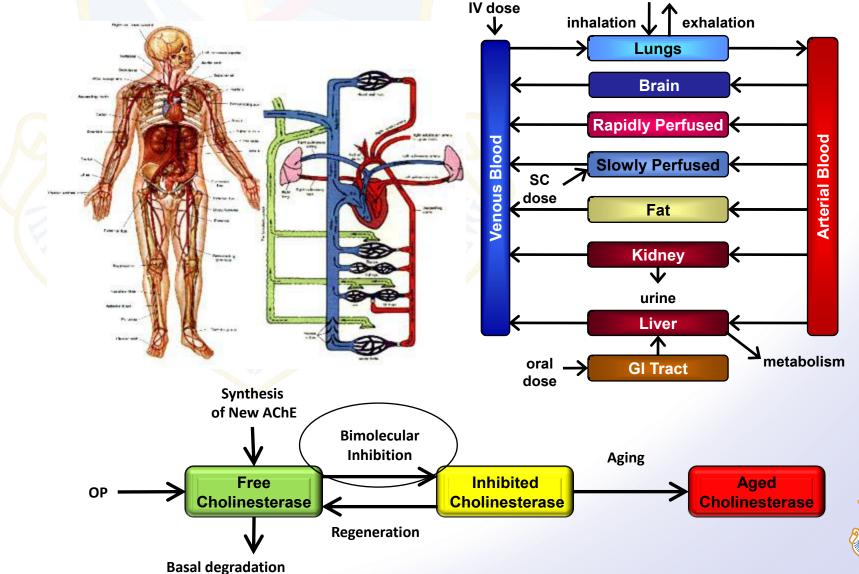
"OP Mechanism of Action"





Physiologically Based Pharmacokinetic/Pharmacodynamic Modeling







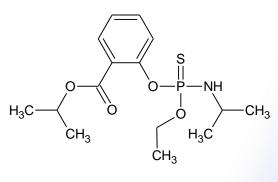
Organophosphate Structures



$$O = \begin{pmatrix} CH_3 \\ O \\ N-P-S \\ O \\ CH_3 \end{pmatrix}$$

CI CI

N
S
II
O-P-O
CH₃
CH₃



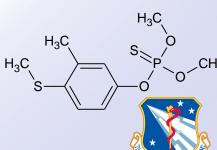
Acephate

Malathion

Chlorpyrifos

Isofenphos

Acetylcholine



Parathion

Glyphosate

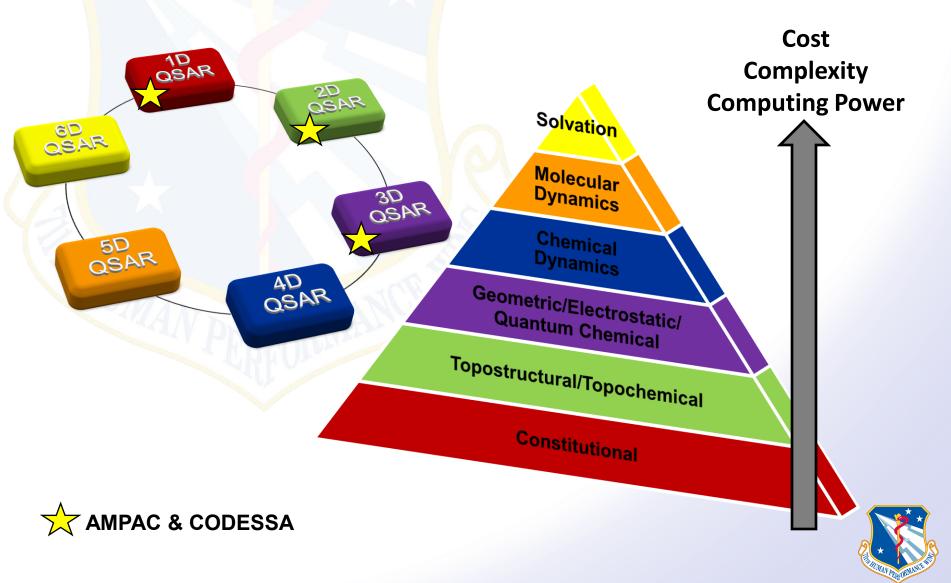
Diazinon

Fenthion



Physiochemical Descriptors







Constitutional Descriptors



Reflect molecular composition of compound without using geometry or electronic structure of molecule:

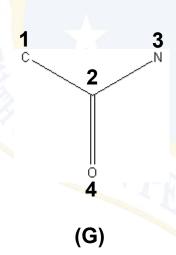
- Number of atoms
 - -Absolute and relative numbers of C, H, O, S, N, F, CI, Br, I, P atoms
- Number of bonds
 - Absolute and relative numbers of single, double, triple and aromatic bonds
- Number of rings
 - Number of rings divided by the number of atoms, number of benzene rings, number of benzene rings divided by the number of atoms
- Molecular and average atomic weight



Topostructural Descriptors



A molecular graph is made up of Edges and Vertices



Adjacency matrix (A)

$$A (G) = \begin{pmatrix} (1) & (2) & (3) & (4) \\ (1) & 0 & 1 & 0 & 0 \\ (2) & 1 & 0 & 1 & 1 \\ (3) & (4) & 0 & 1 & 0 & 0 \\ (4) & 0 & 1 & 0 & 0 \end{pmatrix}$$

Distance matrix (D)

D (G)=
$$(1)$$
 (2) (3) (4)
 (1) 0 1 2 2
 1 0 1 1
 2 1 0 2
 $4)$ 2 1 2 0

Many topostructural indices can be derived from matrices A and D





Regression Techniques



Linear Regression Examples

- Heuristic
- Partial Least Squares (PLS)
- Principle Component Regression (PCR)
- Orthogonal Projection to Latent Structures (OPLS)
- Ridge Regression

Non-Linear Regression Examples

- Support Vector Machines (SVM)
- Neural Networks (NN)
- Kernel Orthogonal Projection to Latent Structures (KOPLS)
- •Kernel Partial Least Squares (KPLS)

Clustering Regression Examples

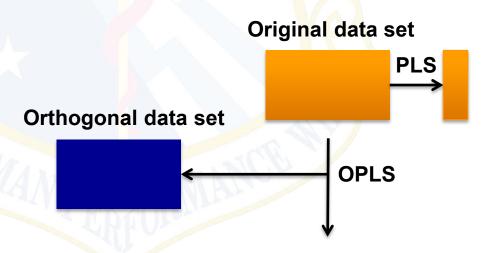
- •k-nearest neighbor
- Random Forest



Regression Techniques



Overview of orthogonal projection to latent structures (O-PLS)



- **Harder to interpret**
- More PLS components
- Orthogonal variation in X

- **Evaluate orthogonal variation** in principal components
- Identify source of orthogonal variation

OPLS treated data



- **Easier to interpret**
- **Fewer components**
- More relevant



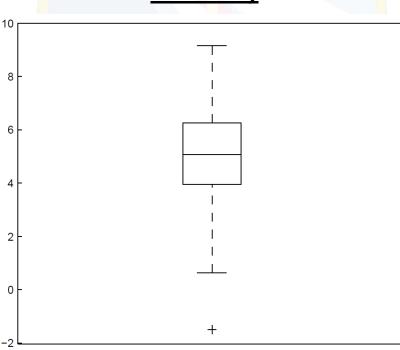


og Bimolecular Rate (M⁻¹min⁻¹)

Acetylcholinesterase Bimolecular Rate Constants (M⁻¹min⁻¹)







OP Compounds

Data collected from 69 peer-reviewed journal articles.

Table 1. Percentage of database that represents a particular temperature and species.

	Species	25	37	Unknown	5	30	22	27	Total Percent
	Human	28.46	6.93	8.15	0.00	0.00	0.00	0.19	43.72
ĺ	Bovine	21.44	3.56	0.09	0.09	1.22	0.00	0.00	26.40
N	Unknown	7.77	0.28	1.50	0.00	0.00	0.00	0.00	9.55
	Fly	2.25	0.56	4.87	0.00	1.03	0.00	0.00	8.71
1000	Rat	0.00	1.22	0.47	0.66	0.00	0.47	0.00	2.81
	Hen	0.56	0.47	0.00	0.00	0.00	0.47	0.00	1.50
	Rabbit	0.47	0.00	0.00	0.00	1.03	0.00	0.00	1.50
Alleri	Eel	0.75	0.75	0.00	0.00	0.00	0.00	0.00	1.50
	Cricket	0.66	0.00	0.00	0.00	0.00	0.00	0.00	0.66
	Guinea Pig	0.19	0.00	0.00	0.00	0.00	0.47	0.00	0.66
	Pig	0.00	0.66	0.00	0.00	0.00	0.00	0.00	0.66
	Mouse	0.19	0.00	0.00	0.00	0.09	0.28	0.00	0.56
	NHP	0.00	0.00	0.00	0.00	0.00	0.47	0.00	0.47
	Catfish	0.00	0.00	0.00	0.00	0.00	0.47	0.00	0.47
	Frog	0.00	0.00	0.00	0.00	0.00	0.47	0.00	0.47
	Minipig	0.00	0.37	0.00	0.00	0.00	0.00	0.00	0.37
	Total Percent	62.73	14.79	15.07	0.75	3.37	3.09	0.19	100.00



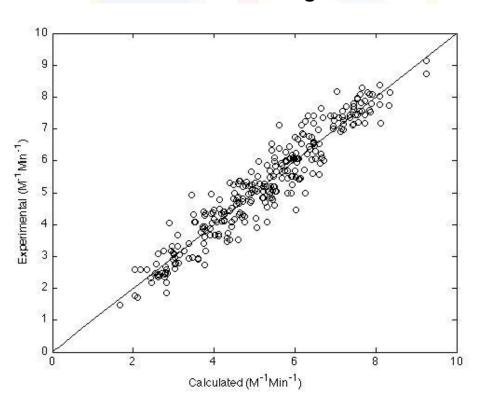




Acetylcholinesterase Bimolecular Rate Constants (M⁻¹min⁻¹)

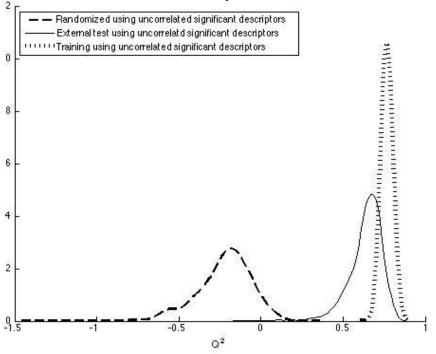


Global Human Orthogonal-PLS



Global training R²=0.91 using 74 significant and uncorrelated descriptors.

Monte Carlo/Bootstrap Cross-Validation "Leave-random-number-out" Consensus QSAR Predictions



A mean training R² of 0.77±0.02 and an external test set Q² of 0.64±0.10 was achieved using the significant uncorrelated descriptors. Y-randomization Q²=-0.23±0.18.

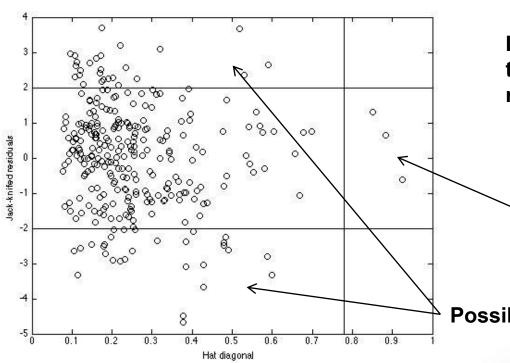




Domain of Applicability



 A number of techniques exist to quantify the Domain of Applicability.



QSAR model predictions are only valid within the applicability domain.

If your test compound falls within the DOA then you can expect a reliable prediction.

Compounds with high leverage can heavily influence a model. Predicted responses outside of the warning leverage may not be reliable.

Possible outliers



AChE Descriptor Significance



Descriptor Name	Normalized P value
Avg nucleoph. react. index for a C atom	1.000
HOMO energy	0.995
Min nucleoph. react. index for a O atom	0.991
Max nucleoph. react. index for a C atom	0.947
Max n-n repulsion for a C-H bond	0.889
(1/6)X GAMMA polarizability (DIP)	0.883
1X GAMMA polarizability (DIP)	0.883
Max e-n attraction for a C-H bond	0.831
HOMO - LUMO energy gap	-0.827
ESP-Max net atomic charge for a F atom	-0.827



Regression Techniques



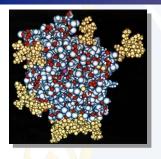
Heuristic (Built into CODESSA 2.51)

- Pre-selection of descriptors based upon a series of criteria cutoffs
 - Variation in descriptors
 - F-test
 - R²
 - T-value
 - Inter-correlation
- F-test measures significance of the whole model, t-test reflects significance of the parameter.

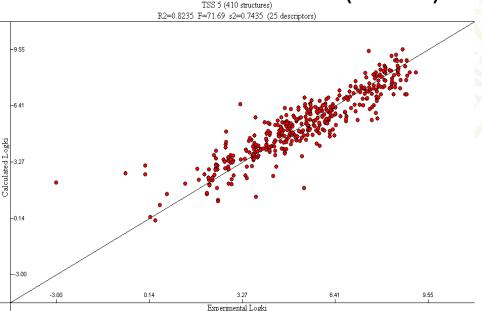


Butyrylcholinesterase "Serum Cholinesterase"



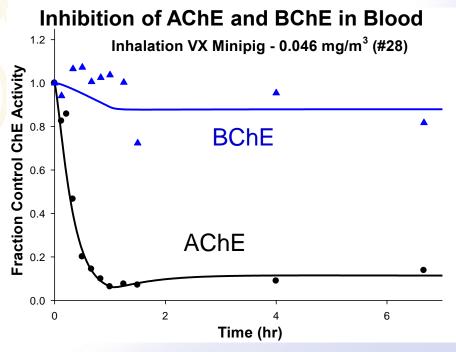


Bimolecular Rate (M⁻¹min⁻¹)



R²=0.82, 25 descriptors, F=71.69, 410 compounds



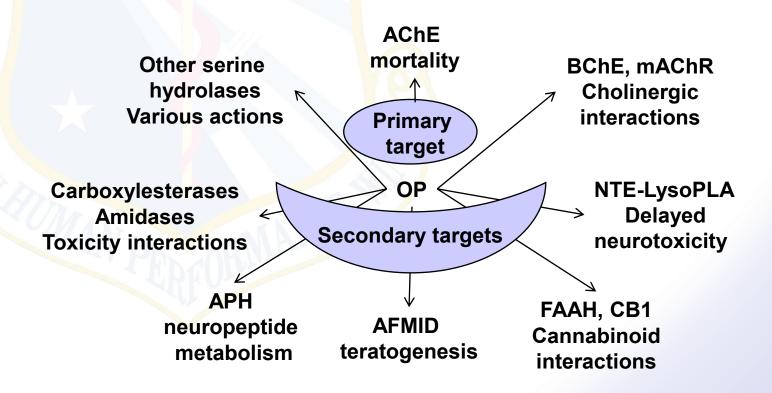


Data taken from literature.



Noncholinergic Targets

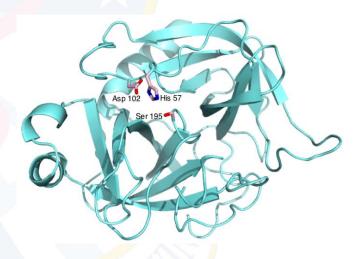




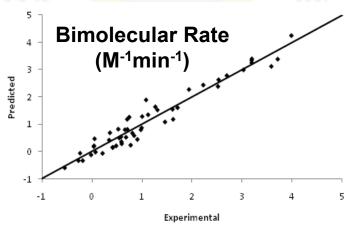


Noncholinergic Targets Digestive Proteases

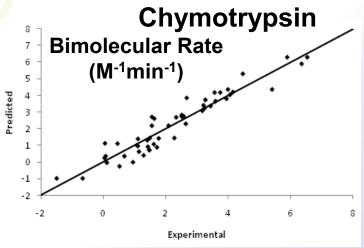




Trypsin



R²=0.94, **Q**²=0.90, **5**2 structures, **1**0 descriptors



R²=0.92, Q²=0.87, 62 structures, 10 descriptors





External Validation



Table 5. Trypsin results from the external validation using the ABC approach.

Training set	Number of compounds	R ²	Q ²	F	s ²	Test set	Number of compounds	R ² _{test}	RMSE _{test}
A + B	35	0.95	0.88	48.46	0.10	С	17	0.85	0.46
A + C	35	0.94	0.91	77.02	0.11	В	17	0.59	0.70
B + C	34	0.91	0.84	30.89	0.15	Α	18	0.82	0.56
Average	34.67	0.93	0.88	52.12	0.12	Average	17.33	0.75	0.57

Table 6. α-Chymotrypsin results from the external validation using the ABC approach.

Training	Number of	R ²	Q ²	F	s ²	Test	Number of	R ² _{test}	RMSE _{test}
Set	Compounds		•	'	3	Set	Compounds	test	TAMOLiest
A + B	42	0.86	0.79	26.24	0.66	С	20	0.86	0.91
A + C	41	0.90	0.63	34.21	0.49	В	21	0.81	1.13
B + C	41	0.68	0.58	14.55	1.36	Α	21	0.16	1.89
Average	41.33	0.81	0.67	25.00	0.84	Average	20.67	0.61	1.31

R²=Coefficient of determination.

Q²=Cross-validated LOO R².

F=Fisher F-test.

 s^2 =Mean squared error. $s^2 = \sum_{i=1}^{Ns} ((Yic - Yio) * (Yic - Yio))/(Ns - Nd - 1)$ where Yic is the ith calculated/predicted property value, Yio is the ith observed/input property value, Ns is the number of training structures, Nd is the number of descriptors and the sum runs from 1 to Ns.

RMSE: Root mean standard error.





Trypsin Descriptors



Descriptor Code	Descriptor Name	T-test (Global training set)	T-test (AB training set)	T-test (AC training set)	T-test (BC training set)
NAª	Error	4.40	3.23	-6.87	0.77
D ₁	Number of F atoms	7.82	6.63	12.65	4.92
D ₂	Kier shape index (order 2)	9.83	8.09	8.12	4.08
D_3	RNCG Relative negative charge (QMNEG/QTMINUS) [Zefirov's PC]	-0.84	0.10	8.33	1.80
D ₄	Kier&Hall index (order 3)	-7.49	-5.06	-6.66	-3.80
D ₅	Balaban index	-3.37	b	-2.65	-3.40
D ₆	PPSA-3 Atomic charge weighted PPSA [Zefirov's PC]	-5.81	-4.75	b	-3.40
D ₇	Number of O atoms	-5.24	-3.92	b	b
D ₈	Relative number of H atoms	-5.11	-5.17	b	1.03
D_{g}	FPSA-1 Fractional PPSA (PPSA-1/TMSA) [Zefirov's PC]	3.66	3.10	b	b
1 D ₁₀	Kier shape index (order 3)	2.00	2.08	-1.23	-0.49



Conclusion



- 1. QSAR can be used to predict organophosphate oxon bimolecular rate constants for AChE, BChE, trypsin and chymotrypsin.
- 2. Approach can be applied to other PBPK/PD modeling parameters.
- 3. QSAR descriptors can provide a mechanistic description of the enzymatic reactions.

Steric hindrance, connectivity, lipophilicity, electrophilicity, electrostatics, hydrogen bonding, van der Waals



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